

RESEARCH ARTICLE

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Key Points:

- WRF with aerosol-aware microphysics is used to study smoke impacts on multiple tornado outbreaks
- Although smoke is always present, changes in environmental conditions due to smoke are highly variable
- WRF and WRF-Chem simulations can produce similar AOD and smoke impacts

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Assessment of biomass burning smoke influence on environmental conditions for multiyear tornado outbreaks by combining aerosol-aware microphysics and fire emission constraints

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Abstract We use the Weather Research and Forecasting (WRF) system to study the impacts of biomass burning smoke from Central America on several tornado outbreaks occurring in the U.S. during spring. The model is configured with an aerosol-aware microphysics parameterization capable of resolving aerosol-cloud-radiation interactions in a cost-efficient way for numerical weather prediction (NWP) applications. Primary aerosol emissions are included, and smoke emissions are constrained using an inverse modeling technique and satellite-based aerosol optical depth observations. Simulations turning on and off fire emissions reveal smoke presence in all tornado outbreaks being studied and show an increase in aerosol number concentrations due to smoke. However, the likelihood of occurrence and intensification of tornadoes is higher due to smoke only in cases where cloud droplet number concentration in low-level clouds increases considerably in a way that modifies the environmental conditions where the tornadoes are formed (shallower cloud bases and higher low-level wind shear). Smoke absorption and vertical extent also play a role, with smoke absorption at cloud-level tending to burn-off clouds and smoke absorption above clouds resulting in an increased capping inversion. Comparing these and WRF-Chem simulations configured with a more complex representation of aerosol size and composition and different optical properties, microphysics, and activation schemes, we find similarities in terms of the simulated aerosol optical depths and aerosol impacts on near-storm environments. This provides reliability on the aerosol-aware microphysics scheme as a less computationally expensive alternative to WRF-Chem for its use in applications such as NWP and cloud-resolving simulations.

1. Introduction

Numerical weather prediction (NWP) is routinely used by forecasters as a tool (among many others) to warn the population of possible severe weather events and is thought to be at least partially responsible for a decrease in death tolls during tornado outbreaks [Brooks and Doswell, 2002]. When building convective outlooks and issuing “tornado watches,” NWP is generally used to forecast the environmental conditions that are favorable for the formation of tornadoes. These conditions include high low-level wind shear, storm-relative helicity (SRH), and convective-available potential energy (CAPE) and low lifting condensation level (LCL, roughly the cloud base height), which are often combined in composite parameters (e.g., Significant Tornado Parameter (STP)) to provide a combined score [Rasmussen and Blanchard, 1998; Thompson et al., 2003]. Recent studies have begun analyzing the influence of large-scale phenomena such as the Madden-Julian Oscillation [Barrett and Gensini, 2013] and climate change [Brooks, 2013; Diffenbaugh et al., 2013] on these precursor atmospheric conditions common to tornado days. The mechanisms through which these environmental conditions are connected to tornado genesis, longevity, and intensity have been hypothesized [Markowski and Richardson, 2009] but remain as open questions. However, as forecasts of these conditions have shown skill on predicting severe weather outbreaks [e.g., Hamill et al., 2005; Knupp et al., 2013], a continuous effort to improve these forecasts further is needed.

Aerosols can interact with clouds and solar radiation and modulate climate [Boucher et al., 2013]. The inclusion of these interactions into NWP can impact forecasts [e.g., Kolsu et al., 2015]. Convective-scale weather

prediction models are moving toward the inclusion of these interactions by the use of aerosol-aware microphysics that incorporate aerosols explicitly in a simple and cost-effective manner [Lebo and Morrison, 2013; Thompson and Eidhammer, 2014]. More complex schemes that resolve atmospheric chemistry and represent the size and composition distributions of aerosols with greater detail [e.g., Eidhammer et al., 2014] are currently too computationally expensive to be used in operational high-resolution forecasts. A comparison between these two types of schemes could be performed to assess how aerosols may impact specific weather phenomena and how much complexity in aerosol treatment may be warranted for future operational NWP.

Biomass burning aerosol (smoke) is estimated to be the major contributor to the global burden of fine carbonaceous aerosols [Andreae and Rosenfeld, 2008; Bond et al., 2004] and to significantly influence climate [Jacobson, 2014]. In particular, smoke from fires in Central America has been shown to be in the inflow of severe weather in the U.S. and has been hypothesized to intensify the outbreaks by the convective invigoration mechanism [Wang et al., 2009]. Saide et al. [2015b] found that Central American smoke was capable of intensifying tornado outbreaks by lowering LCL and increasing low-level shear. This is achieved through two pathways: (1) optical thickening of shallow clouds present before the outbreak which reduce downward solar radiation at the surface stabilizing the boundary layer and (2) enhancement of the capping inversion by heating of the layer above cloud by soot absorption [Saide et al., 2015b]. These studies focused on specific outbreaks and there has been no assessment of these aerosol-associated impacts for multiple outbreaks on multiple years.

This study intends to advance our understanding on the interactions of smoke with severe weather and to help transition the inclusion of fully coupled aerosol-cloud-radiation interactions into NWP. To do so, we compare simple and complex treatments of these interactions in the Weather Research and Forecasting (WRF) and WRF-Chem systems, explore the influence of smoke presence on tornado outbreaks occurring in multiple years, and evaluate the relevance of the proposed mechanism proposed in Saide et al. [2015b] in these outbreaks (section 3). Furthermore, biomass burning emissions have been found to have large uncertainties; for instance, Kaiser et al. [2012] found that a factor of 3 increase in smoke emissions is needed to improve agreement with observations, while Zhang et al. [2014] found discrepancies of up to a factor of 10 between various smoke emission estimates. Thus, to reduce these uncertainties, in this study biomass burning emissions are constrained for each outbreak with satellite-based aerosol optical depth (AOD) retrievals using a recently developed algorithm [Saide et al., 2015a]. The next section describes details on the tornado outbreaks studied, the modeling framework, and the inversion algorithm. Section 3 contains results and discussion about the emission inversions and smoke presence and impacts during outbreaks, while in the last section we provide conclusions and future directions.

2. Methods

2.1. Tornado Outbreaks Studied

We selected eight major tornado outbreaks between 2003 and 2014. This period was chosen due to availability of data from the Moderate Resolution Imaging Spectroradiometer (MODIS) on both Terra and Aqua satellites (operational from mid-2002 to date) as fire emissions and AOD retrievals used herein are based on the MODIS products (see next sections). Only episodes occurring during April and May (not including those very early and very late in each month, respectively) were chosen to coincide with the strongest biomass burning in Central America [Reid et al., 2004]. The tornado tracks of the outbreaks selected are shown in Figure 1 and represent locations across the southeast, midwest, and Southwest United States (U.S.). Other selection criteria were for each outbreak to contain at least four tornadoes of EF2 scale or greater (to filter out the less significant outbreaks), the tornadoes were not so broadly spread in order that model simulations would fit reasonably within a nested domain, and the most severe and more numerous tornadoes needed to occur in the afternoon or early evening. The latter condition is imposed because the proposed mechanisms being studied are driven by peak solar heating, which would dissipate at night or not yet occur in the morning. Many of these outbreaks have been highlighted as major events of the year they occurred [Blunden and Arndt, 2012, 2014, 2015; Levinson and Lawrimore, 2008; Levinson and Waple, 2004; Peterson and Baringer, 2009], and specific studies have focused on their characteristics and impacts [Hamill et al., 2005; Knupp et al., 2013]. The environmental conditions provided by Storm Prediction Center (SPC) Convective Outlooks and Mesoscale Analysis for these outbreaks are shown in Figure 2.

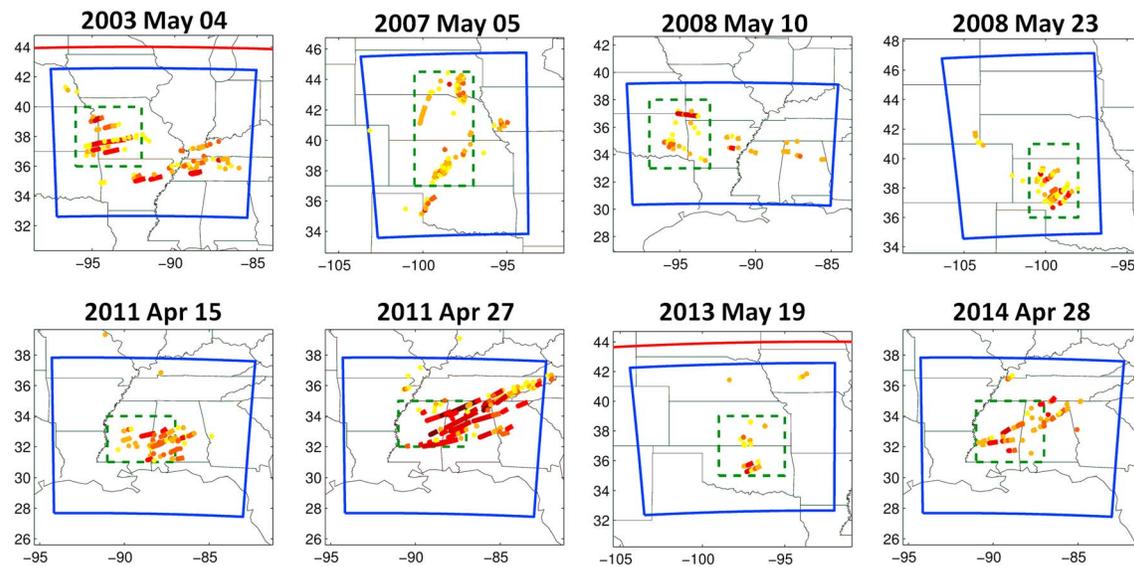


Figure 1. Tornado tracks of the outbreaks studied color coded by Enhanced Fujita (EF) Scale [Potter, 2007], with darker colors representing higher magnitude. Blue solid lines represent the inner domain specified for each simulation. Boundaries of the outer domain are in red solid lines when included in the regions shown (see full outer domains in Figure 2). Green segmented lines show the region where statistics are computed which corresponds to the area where the earlier tornadoes occur.

2.2. Regional Modeling

We use two configurations of the WRF system to perform simulations of aerosols and their impacts on weather, one with the *Thompson and Eidhammer* [2014; hereafter referred to as TE2014] aerosol-aware microphysics (AAM) scheme and the other with WRF-Chem.

The TE2014 aerosol-aware microphysics consists of a double-moment bulk microphysical parameterization that explicitly resolves droplet nucleation and ice activation due to aerosols. Besides cloud water, cloud ice, snow, graupel, and rain hydrometeor species, the scheme transports two aerosol species (hygroscopic and ice nucleating) adding only about 15% computational cost, therefore making it suitable for NWP applications. Aerosol-radiation interactions were recently included to the AAM configuration by computing AOD at 550 nm based on the two aerosol species using a lookup table procedure to include aerosol hygroscopic growth. Then, spectral AOD and other aerosol optical properties (AOPs) such as Angstrom exponent, single-scattering albedo, and asymmetry factor are parameterized following the methods of *Ruiz-Arias et al.* [2014] considering a rural-type aerosol. These spectral AOPs are then used in the rapid radiative transfer model (RRTMG) short-wave radiation parameterization [*Jacono et al.*, 2008] to include aerosol effects on these calculations.

While TE2014 used a constant flux derived from initial aerosol concentrations to represent emission processes, we modified WRF-Chem emission routines to include primary-aerosol number emissions explicitly into the aerosol-aware microphysics parameterization. The Model for Simulating Aerosol Interactions and Chemistry (MOSAIC) aerosol scheme [*Zaveri et al.*, 2008] in WRF-Chem can solve aerosol mass and number for eight sectional size bins. Thus, emission routines coupled to this scheme (anthropogenic, fires, dust, and sea salt) provide aerosol number per size bin, which were summed up to obtain total aerosol number concentrations needed by the TE2014 scheme. We used biomass burning emissions produced by the Quick Fire Emission Data set (QFED) v2.4 [*Darmenov and da Silva*, 2015] which were included into the model using the WRF-Chem online plume rise model [*Grell et al.*, 2011]. Anthropogenic emissions correspond to NEI 2005 (<http://www.epa.gov/ttnchie1/net/2005inventory.html>) for the U.S., 1999 Mexico emissions described in *Mena-Carrasco et al.* [2009] updated to 2012 using growth factors from *Wolf et al.* [2009], and emissions derived with PREP-CHEM-SRC code [*Freitas et al.*, 2011] for the rest of Central America. Sea salt and dust emissions were estimated online by using the *Gong et al.* [1997] and *Zhao et al.* [2010] parameterizations, respectively. Although secondary aerosol production is not modeled (no chemical mechanism used and no gaseous species modeled), it should not affect aerosol number concentration significantly as secondary aerosol is expected to condense on already existing particles. Aerosol boundary conditions are provided every 6 h by global simulations of Model for Ozone and Related chemical Tracers (MOZART) [*Emmons et al.*, 2010], while

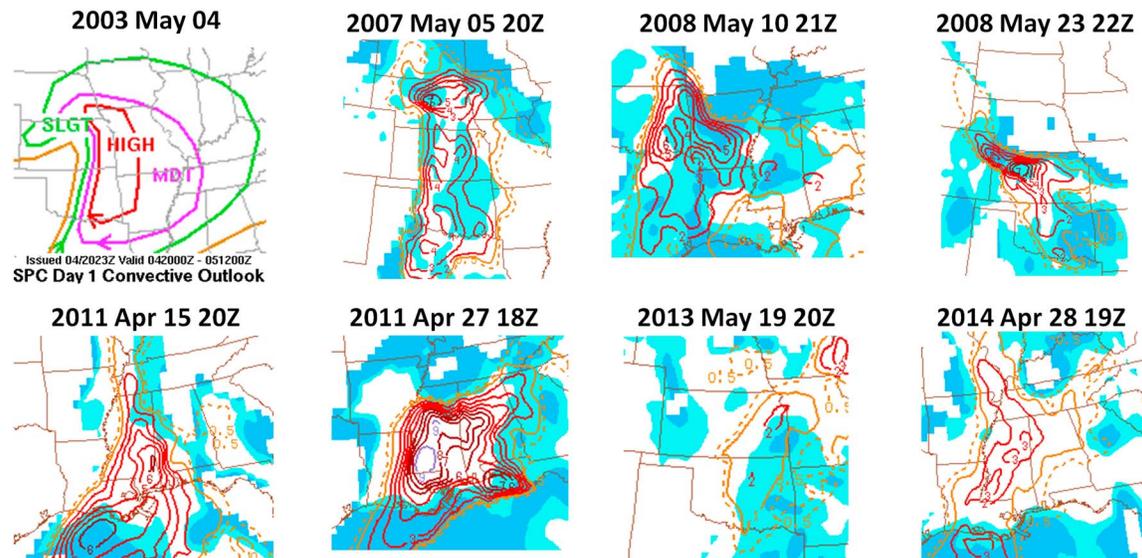


Figure 2. Environmental conditions for the outbreaks studied as provided by the SPC. While the day 1 convective outlook valid at 23 UTC is shown for 4 May 2003, the fixed-layer Significant Tornado Parameter (STP, in contours) and the Convective Inhibition (CIN, shaded at 25 and 100 J/kg) obtained from the hourly Mesoscale Analysis are shown for the rest of the outbreaks.

initial conditions are obtained from the monthly climatology derived by *Thompson and Eidhammer* [2014]. The month when the simulation was started was used as initial condition when fire emissions were included, while February was used for simulations without fire emissions to provide a smoke-free initialization. As in TE2014, only dust was considered to be ice nucleating, while all other aerosol species (except black carbon) were apportioned to the hygroscopic aerosol. Although black carbon is not apportioned, the two aerosol species are considered to absorb solar radiation as mentioned above.

WRF-AAM was configured for each tornado outbreak by using two domains of 12 km and 4 km grid spacing. The outer domains covered roughly between -110° and -75° longitude for all outbreaks and from 10° to 50° latitude for outbreaks in 2007 and 24 May 2008 and from 10° to 45° latitude for the rest of the outbreaks (Figure 3). The inner domains are shown in Figure 1 and were designed to include each outbreak region. All simulations were initialized at 00 UTC 7 days prior to the outbreak studied to provide spin-up time for aerosol concentration and feedbacks to meteorology. Meteorological initial and boundary conditions were obtained from National Centers for Environmental Prediction (NCEP) final analysis [NCEP, 2000]. Other WRF configuration choices include RRTMG long-wave radiation [Iacono *et al.*, 2000], Mellor–Yamada–Janjic (MYJ) boundary layer [Janjić, 2002], and Noah land surface model [Barlage *et al.*, 2010]. The GF cumulus scheme [Grell and Freitas, 2013] is used for the 12 km domain, and no convective parameterization is used for the 4 km domain. Figure 2 compared to Figure 4 shows that the model configured this way is able to represent the tornadic environment found on all outbreaks, although some shifts in the maximum values are often found which is expected due to the long spin-up used.

The WRF-Chem configuration used is the same as the one described in *Saide et al.* [2015b] unless noted differently here. For consistency between simulations, WRF-Chem uses the same configuration as WRF-AMM, including domains, vertical and horizontal resolution, emission databases, initial and boundary conditions, and physics configuration, excluding the microphysics and aerosol activation parameterizations where the Morrison and Abdul-Razak and Ghan schemes are used [Abdul-Razak and Ghan, 2002; Yang *et al.*, 2011, and references therein]. The chemical mechanism and aerosol model correspond to the modified Carbon Bond Mechanism (CBM-Z) [Zaveri and Peters, 1999] and eight size bin MOSAIC [Zaveri *et al.*, 2008].

2.3. Inversion Scheme

The algorithm described in *Saide et al.* [2015a] is used to constrain biomass burning emissions for each outbreak studied. The algorithm requires the use of two WRF simulations, one with first-guess (or initial) emissions (QFED in this case) and another one with perturbed emissions. As assessing smoke impacts involves

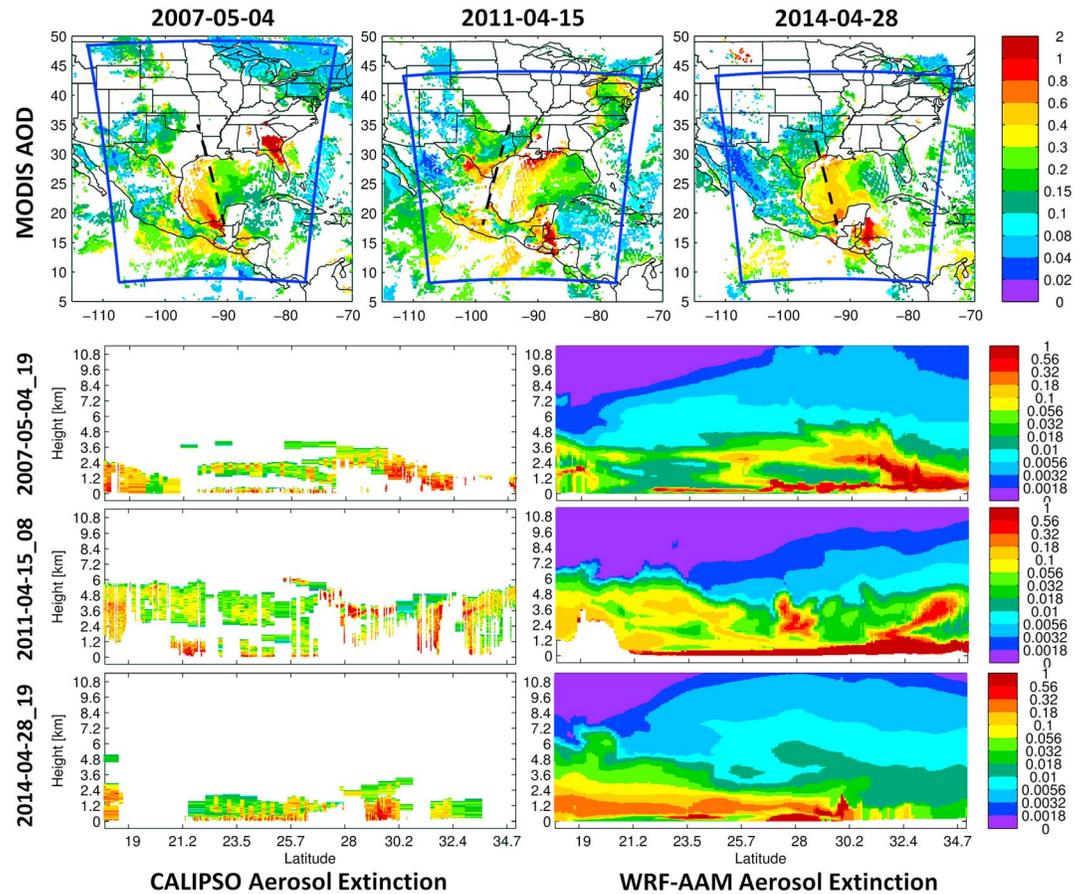


Figure 3. (top row) Observed AOD maps on the day or the day before of three outbreaks studied (5 May 2007 on the left, 15 April 2011 on the middle, and 28 April 2014 on the right) by combining Terra and Aqua overpasses. The solid blue line represents the outer domains used for these outbreaks. (bottom rows) Observed and modeled (Fire ON + inversion) extinction (1/km) curtains. The CALIPSO tracks corresponding to these observations are shown as dashed black lines in Figure 3 (top row).

performing simulations with and without fire emissions [Saide *et al.*, 2015b], the perturbed simulation is chosen as the one with the fire emissions turned off (i.e., perturbation factor equals zero). This reduces the computational burden because the perturbation simulation is used both for the inversion step and for assessing the smoke effects. Also, the simulations need to produce estimates of the observation used to constrain emissions, which in this case is AOD and can indeed be estimated by WRF-AAM (see previous section). Additionally, the simulations used in the inversion algorithm need to solve for tracers tagged to the emissions being constrained. These tracers provide the emission's footprint in the observation location and are used to derive the sensitivities (derivatives) used in the inversion scheme. In this study the tracers are tagged to fire emissions from four regions, every 6 h and during 8 day simulations, resulting in 128 tracers. The four regions are shown in Figure 5 and correspond roughly to the (1) east and (2) west of the Mexican Central Plateau, (3) the Yucatan peninsula including Belize and Guatemala, and (4) the rest of Central American countries to the south. The tracers are modeled only on the 12 km domain as here is where the inversion is performed.

The variational inversion algorithm determines the optimal scaling factors to be applied to the fire emissions. A cost function with two terms is minimized to (1) improve model fit to the observations but at the same time (2) do not exceedingly deviate from the initial guess estimate. We use the same parameters as in Saide *et al.* [2015a] to weight these two terms with the exception of the length scale to correlate emissions temporally, which is changed from 4 to 6 h. This choice of parameters provides a good fit to the data with little bias and produces large deviations from the highly uncertain guess emissions but does not generate very steep temporal changes in the correction factors. Finally, one more simulation is performed with the constrained emissions, and this simulation is compared to the one without fire emissions to assess the smoke impacts. These

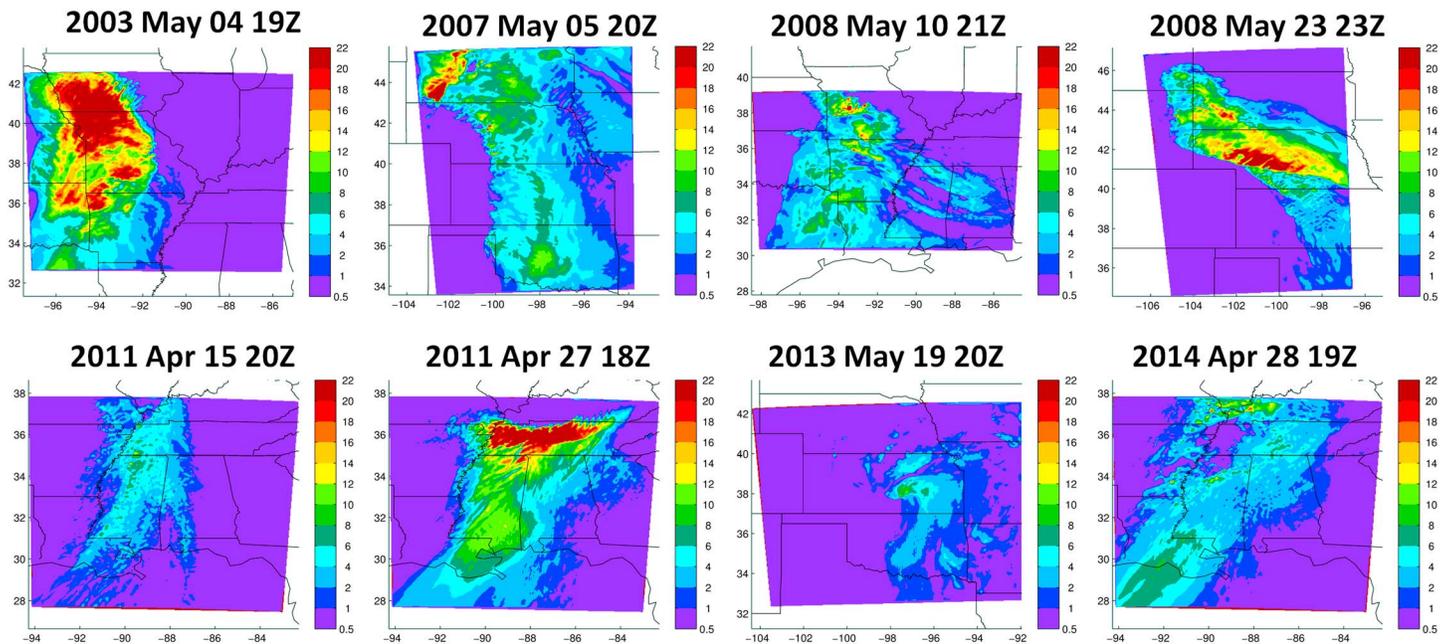


Figure 4. Significant Tornado Parameter (STP) at the beginning of each outbreak studied as modeled by the WRF-AAM configuration when using the constrained fire emissions.

two simulations (constrained fire emissions and no fires) use the two domains (12 km and 4 km resolution), and comparisons are performed on the inner domain only. The inversions are performed only for the WRF-AAM configurations, while WRF-Chem configurations using constrained smoke emissions use the ones obtained from WRF-AAM inversions corresponding to the same episode.

We use observed AOD at 550 nm from the NASA neural-network retrieval (NNR) [NASA Global Modeling and Assimilation Office (GMAO), 2014; Saide *et al.*, 2013] to constrain emissions, which are assimilation-grade retrievals based on MODIS products calibrated to match Aerosol Robotic Network (AERONET) AOD data [Holben *et al.*, 2001]. Only NNR observations from -110° to -80° longitude, from 10° to 30° latitude, and over 0.2 AOD are used in the inversion to avoid correcting fire emissions based on retrievals far from the source region and not significantly influenced by fires.

2.4. Summary of Simulations

We performed three WRF-AAM simulations for the eight outbreaks studied: (1) fire emissions turned off and two domains, (2) initial fire emissions including tracers and only for the 12 km domain, and (3) constrained emissions for the two domains. These sum up to 24 simulations. Then, we performed various sensitivity simulations by turning off absorption and changing initial smoke emissions (see section 3.4 for details) for three outbreaks. One simulation per outbreak was performed for absorption effects (only #3 as fire off was already performed), while two for the initial emissions sensitivity (#2 and #3 as the inversion needs to be repeated). Together with the baseline simulations, this gives a total of 33 WRF-AAM simulations. We also performed simulations #1 and #3 using WRF-Chem for three outbreaks, thus a total of six WRF-Chem simulations (section 3.5). Additionally, one WRF-Chem simulation of type #2 but without the tracers was performed to compare AOD against WRF-AAM. Note that each WRF-Chem simulation is ~ 10 times more computationally expensive than the corresponding WRF-AAM.

3. Results and Discussion

3.1. Inversion Results

Figure 5 shows the daily emissions by region before and after the inversion for all outbreaks studied when using the AAM configuration. Prior smoke emissions are found to be underestimated for most times during all cases studied, with domain-wide emission scaling factors in the range of 1.4–2.6, depending on the case. The corrections to the prior emissions tend to vary substantially by region and by day, which shows that the

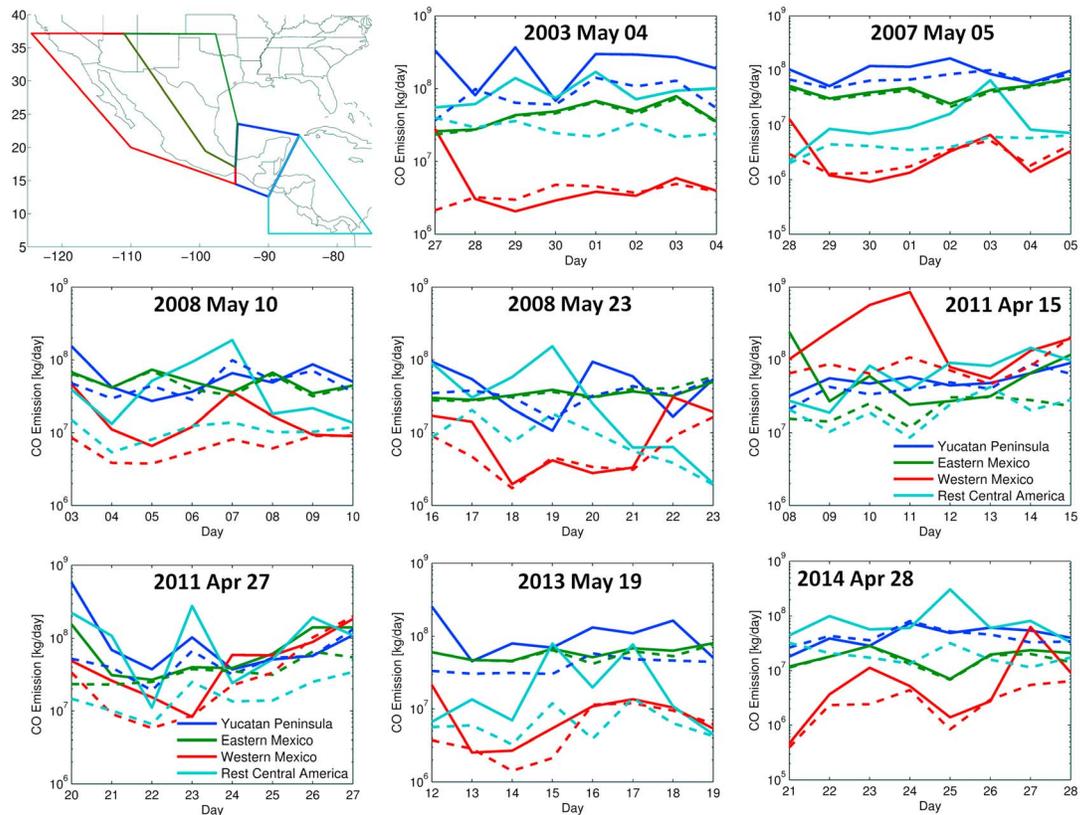


Figure 5. (top left) Regions used to tag fire emissions tracers (see section 2.3). Rest of the panels: Constrained (solid lines) and first-guess (segmented lines) emissions by day on each region and for each outbreak (indicated at the top of each panel).

choice of independently constraining emissions by region and by time was appropriate. Figure 6 shows an example of how the modeled AOD generally improves the fit with respect to the observations after constraining emissions, indicating that the inversion algorithm is functioning properly. Although there is overall improvement, some regions show degraded AOD after the inversion (e.g., southwest Mexico), which can happen because of the limited number of parameters being optimized (four regions, four times a day) and the large amount of data being ingested (i.e., the system is over determined).

3.2. Smoke Presence Within Outbreak Regions

Figure 7 (top row) shows the vertical distributions of aerosol number concentration in the region of each outbreak studied, while Figure 8 shows the boundary layer height distributions. We find that smoke contributes significantly to the total aerosol number concentration for all of the outbreaks analyzed no matter the location of the outbreak. Smoke can be effectively transported to the regions where outbreaks occur due to the southerly airflow from the Gulf of Mexico during spring [Wang *et al.*, 2006, 2009] and because of the low-level jet that transports warm, moist air from the same region, which is a typical feature that triggers these outbreaks [Hamill *et al.*, 2005; Knupp *et al.*, 2013]. The vertical location of smoke can vary from case to case, with all outbreaks showing smoke within the boundary layer and only some cases (15 and 27 April 2011 and 28 April 2014) presenting a thick smoke layer extending higher into the lower troposphere (up to ~5 km). WRF-Chem simulations for the 27 April 2011 outbreak also showed a thick layer of smoke [Saide *et al.*, 2015b], which is consistent with the WRF-AAM results and as expected because both model configurations use the same transport scheme.

CALIPSO provides observations of vertically resolved aerosols, which can be used to evaluate the modeled vertical distribution of the smoke layer when the satellite overpasses it. Since the regions where the outbreaks occur are generally cloudy, we performed the evaluation over the Gulf of Mexico, which is the inflow region (Figure 3). We find the model generally reproduces the layers observed by CALIPSO with skill and at a

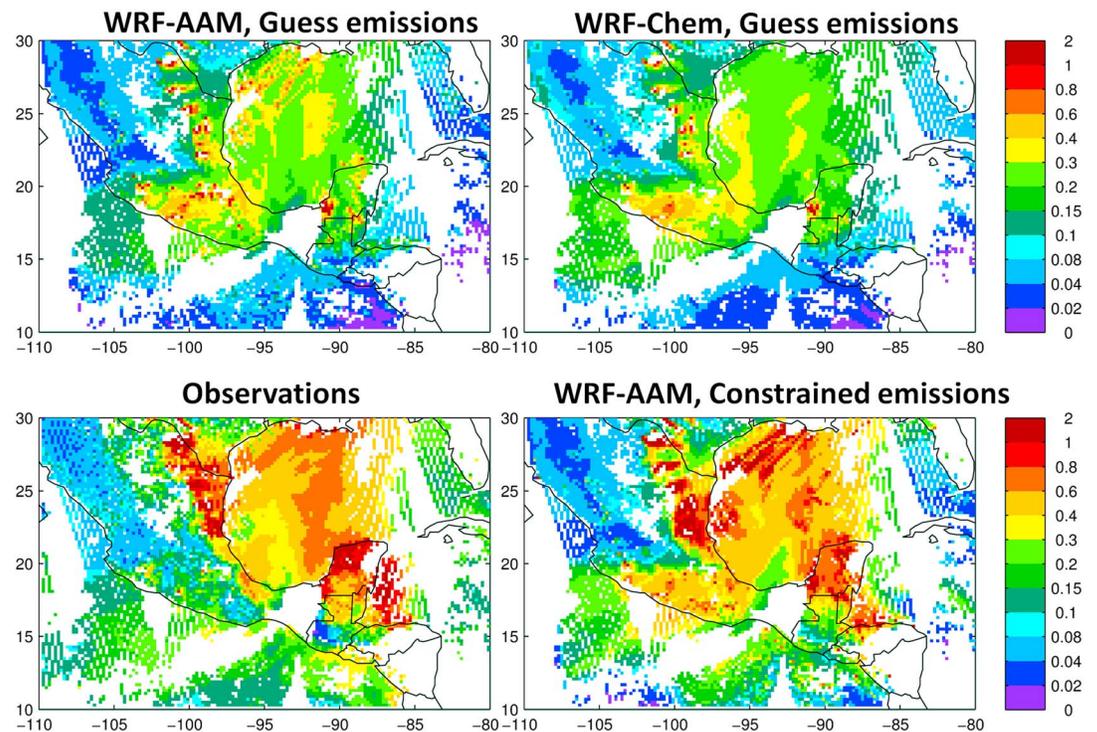


Figure 6. AOD maps for the observations and three model simulations on 27 April 2011 by combining Terra and Aqua overpasses. Cloudy grid cells are not considered when computing model AOD.

similar altitude, going from shallow layers of smoke found on 28 April 2014 to layers up to ~6 km altitude on 15 April 2011. This shows that the plume rise parameterization and the model transport are reliable, which increases the confidence on the vertical distribution of smoke on the outbreaks regions shown on Figure 7.

3.3. Impacts on Environmental Conditions

The changes produced by the presence of smoke in near-storm environment conditions can be obtained by comparing simulations with and without smoke emissions (Figure 9). Although smoke is present in the outbreak region for all cases studied, the impacts of smoke on environmental conditions vary by case. The outbreaks on 4 May 2003 and 27 April 2011 show the largest and most consistent effects of smoke on LCL (~90–130 m reduction in mean), low-level shear (1.5–2.2 m/s increase in mean), and SRH (30–80 m²/s² increase in mean), with the interquartile region showing values that produce an intensification of tornadic environments (negative LCL and positive 1 km shear and SRH differences). Statistics of these variables for multiple events have shown that differences in median between supercell types (nontornadic to weakly tornadic and weakly to significantly tornadic) are ~ -170 m, 1.7 m/s, and ~30–45 m²/s² for LCL, low-level shear, and SRH, respectively [Thompson *et al.*, 2003]. Thus, the smoke effects shown here can reach magnitudes relevant for tornadogenesis and increase in tornado intensity.

Given that the statistics in Figure 9 are computed using a large number of grid cells (7000–23,000 depending on the case), statistically significant differences are likely to be found even if differences between simulations are small. Thus, instead of using *p* values, we estimate the magnitude of the difference between groups (effect size) using the standardized difference (difference between means divided by the standard deviation of either group) and consider small, medium, large, and very large values of this quantity equal to 0.2, 0.5, 0.8, and 1.3, respectively [Sullivan and Feinn, 2012]. As expected, the outbreaks on 4 May 2003 and 27 April 2011 show medium to large effect size for LCL, 1 km shear and SRH (Figure 9). While 10 May 2008, 15 April 2011, and 28 April 2014 generally show a small effect size, the rest of the outbreaks show negligible effects of smoke. Although the 23 May 2008 and 19 May 2013 outbreaks show positive LCL mean (i.e., a reduction in tornado likelihood due to smoke), this is found to be a negligible effect size due to the large spread in these simulations.

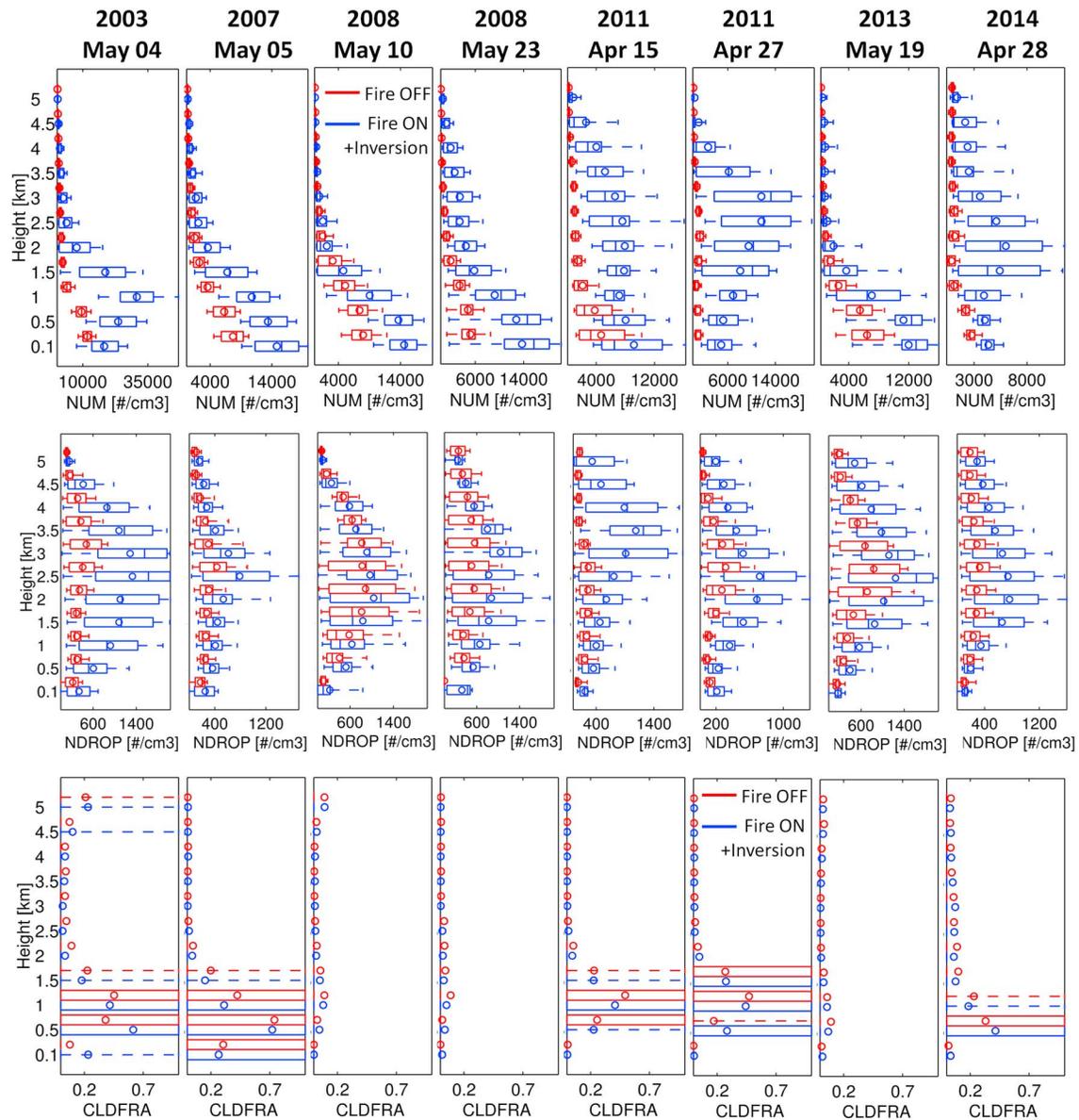


Figure 7. Vertical distributions of modeled (WRF-AAM) aerosol number concentration (NUM), cloud droplet concentration (NDROP), and cloud fraction (CLDFRA) for the cases studied turning on and off biomass burning emissions. The distributions are shown as box plots, with center solid lines indicate the median, circles representing the mean, boxes indicating upper and lower quartiles, and whiskers showing the upper and lower deciles. Statistics are computed for the regions shown in Figure 1 and 1 h before the outbreak starts. Note that cloud fraction is either 0 or 1 in this WRF configuration; thus, the boxes and whiskers either use the whole 0–1 interval or collapse to zero for low cloud fractions. X axes start at a value of 0 for all panels.

Elevated CAPE and large-scale shear (6 km layer) are generally understood as a requirement for supercell formation rather than an indicator to discriminate between tornadic categories [Thompson et al., 2002, 2003]. Figure 9 shows the effects of smoke on these two variables. For CAPE, although the outbreak on 4 May 2003 shows a very large increase in CAPE due to smoke, the simulation without fires has mean CAPE of ~1500 J/kg which is high enough to support the formation of supercells in the model (not shown). On the other hand, for the 27 April 2011 outbreak there is only a medium effect size showing a decrease in CAPE due to smoke, but again, both simulations show mean CAPE of over 2000 J/kg; thus, this difference is not expected to impact supercell formation. On the other hand, all cases show a negligible effect of smoke on 6 km shear, which is different to the effects mentioned earlier for 1 km shear. This indicates that changes in wind due to smoke are mostly located in the lower layers. Figure 9 also shows convective inhibition (CIN), which is also a parameter used to assess the formation of supercells, indicating a small to negligible effect

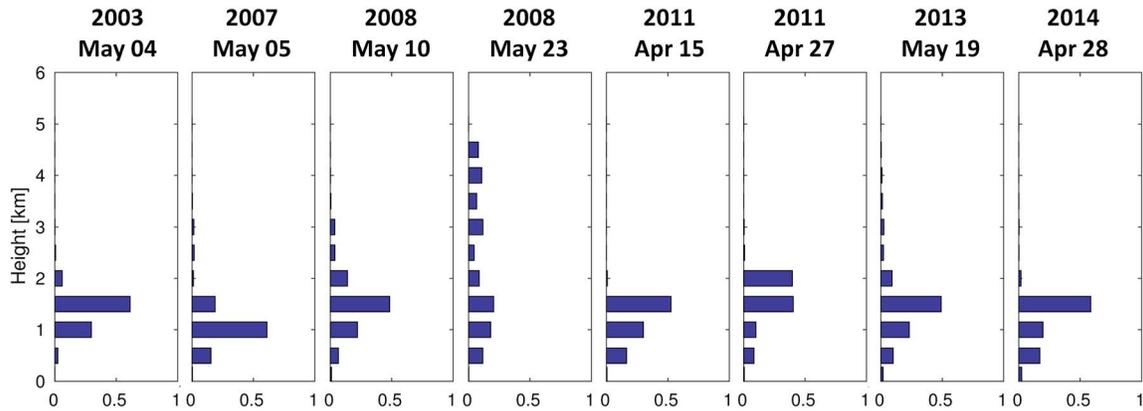


Figure 8. Normalized histograms of boundary layer height for the same regions, height categories, and outbreaks as in Figure 7.

of smoke on this parameter. Although the smoke produces stronger CIN (i.e., more negative) for the 4 May 2003 outbreak, supercells are still generated in regions of close to zero CIN for the simulation including smoke emissions.

Why is there large variability in the smoke effects on environmental conditions although all outbreaks studied consistently show smoke presence? Figure 7 shows cloud droplet number concentrations (NDROP, middle row) and cloud fraction (bottom row) that can help answer this question. For instance, 4 May 2003

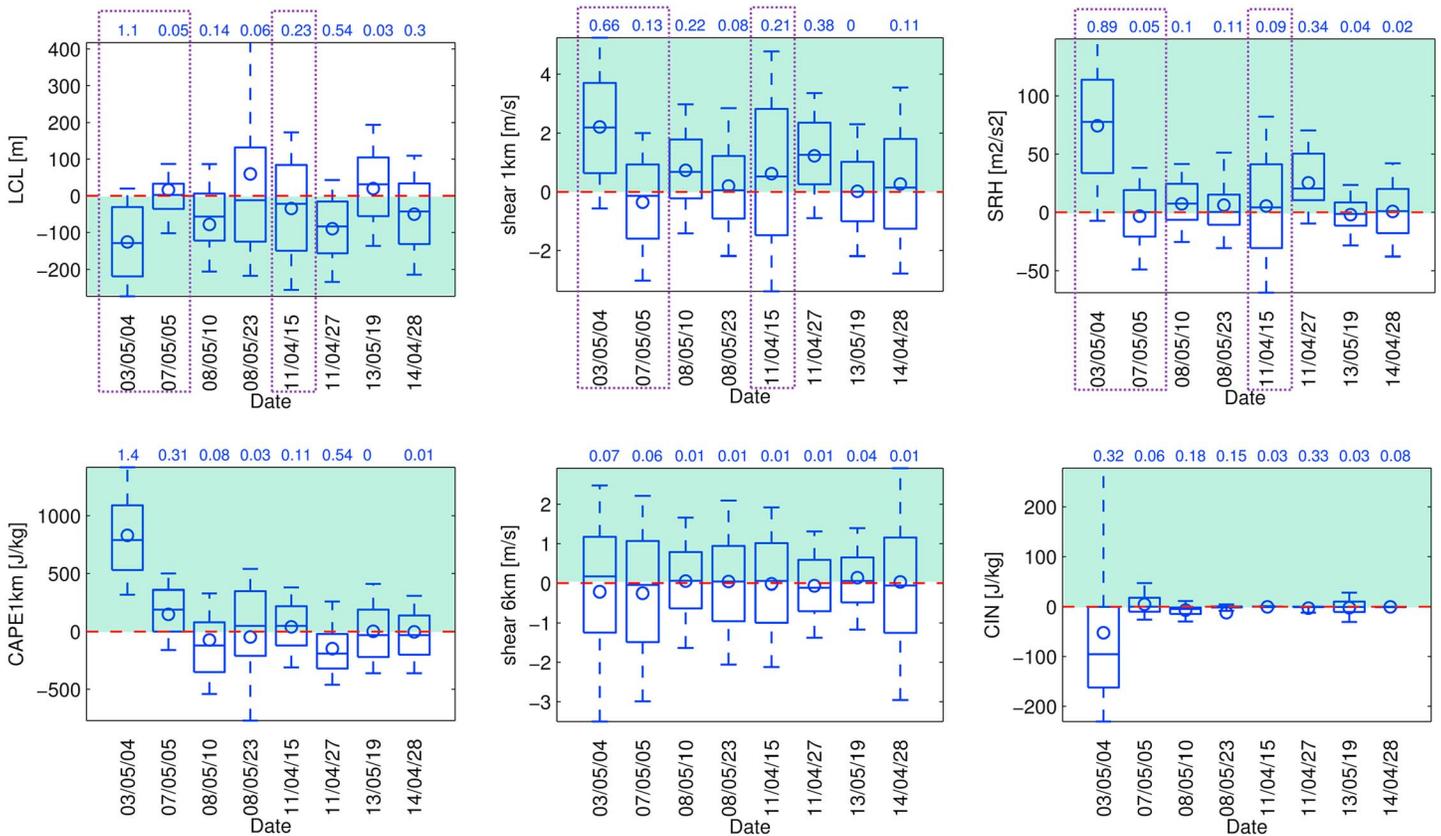


Figure 9. Statistics (as in Figure 7) of the differences between WRF-AAM simulations with and without fire emissions. The standardized difference (effect size) between both simulations is shown on top of each panel. The dashed red lines indicate the zero difference line, while the green area denotes the region where the smoke increases the likelihood of tornado occurrence and intensity (decrease in LCL, increase in 1 km shear and SRH) and a higher chance of convection (increased CAPE, 6 km shear, and CIN). The dotted purple rectangles highlight the sensitivity cases and variables studied in the next figures. Note that CIN is a negative quantity. Dates are in YY/MM/DD format.

and 5 May 2007 show similar patterns of smoke (mostly contained in the boundary layer) and similar cloud fractions (ranging from 0.3 to 0.7 in the 0.25–1.25 km layer). However, for 4 May 2003 the impacts of smoke on NDROP in the 0.25–1.25 km layer are large (twofold to threefold increase on average), while they are small for the 5 May 2007 outbreak (50–67% increase on average), which coincides with a large and negligible effect on environmental parameters for the 4 May 2003 and 5 May 2007 outbreaks, respectively. One of the mechanisms that generate conditions where tornado formation is more likely is the optical thickening of low-level clouds through the activation of additional cloud droplets by the smoke [Saide *et al.*, 2015b]. Thus, if this additional increase in NDROP is not substantial, it is expected that environmental conditions would not change through this mechanism. The 27 April 2011 outbreak, which is the other outbreak with strong smoke effects, also shows a factor of 2–3 increase on average NDROP for vertical layers with cloud fractions over 0.2. Also, this outbreak presents a thick layer of smoke on top of the clouds that is expected to exacerbate the smoke effects by enhancing the capping inversion through soot absorption [Saide *et al.*, 2015b], which in the WRF-AAM configuration is included through the single-scattering albedo (SSA) of the mixture. The rest of the outbreaks either did not present cloud fractions large enough nor sufficiently large changes to the cloud droplet number concentration to produce regional effects on the environmental conditions.

Now, why does cloud droplet concentration increase due to smoke show large variability between outbreaks? Considering only the layers with cloud fractions higher than 0.2, the outbreaks with the largest NDROP increases (4 May 2003 and 27 April 2011) are those with the largest increases in aerosol number concentrations due to smoke (Figure 7). For instance, in the 250–750 m layer there is a threefold to sixfold increase (on average) on aerosol number concentration for the outbreaks with larger NDROP changes versus an approximately twofold increase for the rest of the outbreaks, while for the 0.75–1.25 km layer the increases are sevenfold to ninefold versus threefold to fourfold. Thus, larger aerosol number concentrations due to smoke in the cloudy layers can contribute to intensification of the outbreaks (i.e., lower LCL, larger 1 km shear, and SRH), but it also depends on the overlying background conditions. For instance, the 5 May 2007 outbreak presents larger aerosol number concentrations at the cloud level compared to the 27 April 2011, but since the simulation without fires shows very low aerosol number concentrations for 27 April 2011 the smoke effects are amplified in this event. The vertical location of the smoke and its ability to reach the cloud layer is also crucial, as for instance, the 28 April 2014 outbreak shows fivefold to sixfold enhancements in aerosol number concentration due to smoke above the cloud layer but was not mixed downward to provide large enhancements in NDROP. Other conditions affecting aerosol activation including temperature and vertical wind velocities [Thompson and Eidhammer, 2014] can change from outbreak to outbreak and could also play a role on the differences found.

3.4. Sensitivity Simulations

We performed additional simulations to assess the effects of aerosol absorption and of changing the initial smoke emissions. In the case of absorption, we modified the WRF-AAM code to use single-scattering albedo (SSA) of a marine-type aerosol which absorbs radiation very weakly (SSA of 0.99 at 70% relative humidity versus 0.95 SSA of the rural type aerosol used in the base simulations). The effect of absorption is then obtained by comparing simulations with smoke emissions using the base configuration and the configuration just described. The other set of sensitivity simulations performed used the Fire INventory from NCAR (FINN) [Wiedinmyer *et al.*, 2011] biomass burning emissions as initial estimates and followed the same emission inversion scheme. All these simulations were performed for three tornado outbreaks: 4 May 2003, 5 May 2007, and 15 April 2011. These outbreaks were chosen because they show a range of vertical smoke distributions (mostly in the boundary layer and in the lower troposphere) and smoke effects (from large to negligible effect size).

Figure 10 (top row) shows the effects of smoke on tornado parameters due to absorption. The effect of absorption can vary from negligible to medium effect size depending on the outbreak and the variable. The outbreak on 15 April 2011 presents a thick aerosol layer on top of the shallow clouds going up to 4–5 km, while the smoke layer is mostly at cloud level or below on the other two outbreaks (Figure 11, top left). Previous studies have found that the location of the smoke (or absorbing aerosol) with respect to clouds can be critical for the resulting effects, with above cloud aerosol tending to increase cloudiness by enhancing the capping inversion, while in-cloud aerosol can have the ability of burn-off clouds

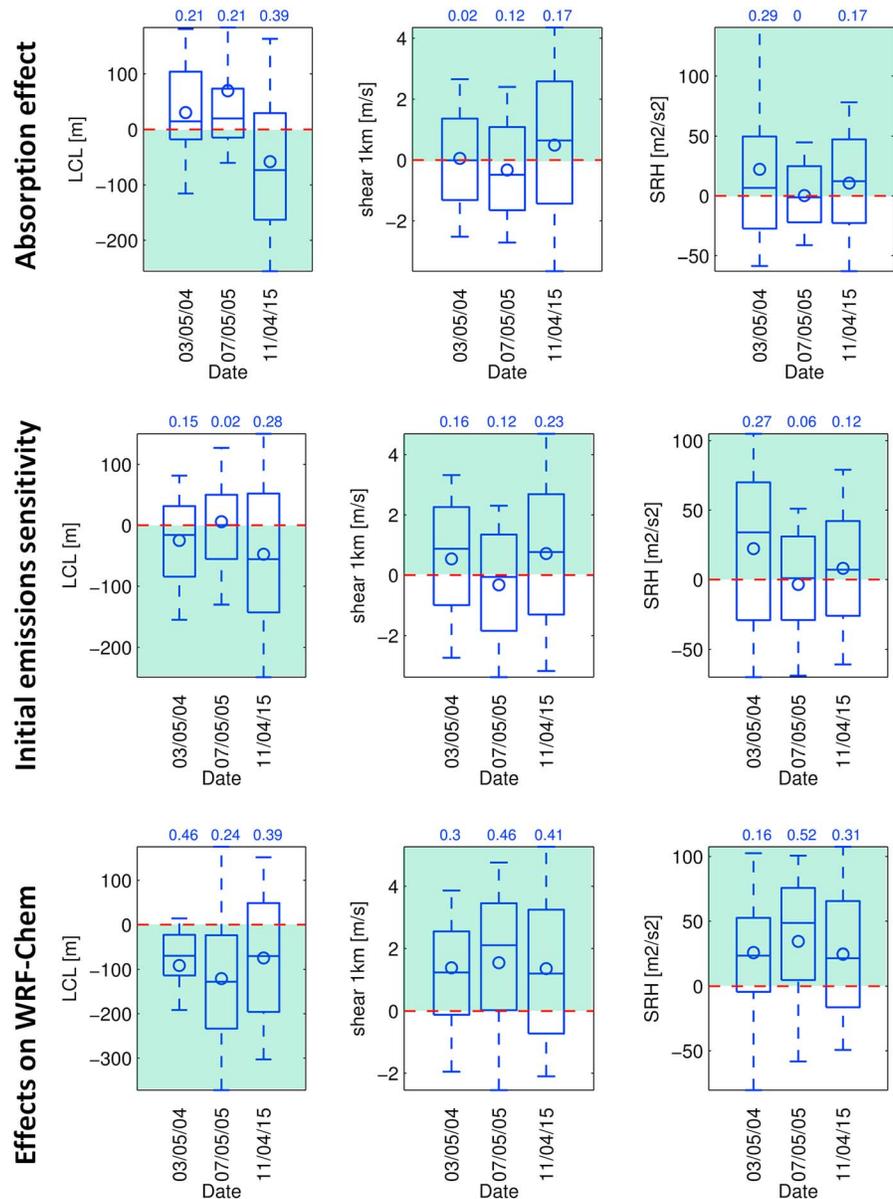


Figure 10. As Figure 9 but for the sensitivity cases and WRF-Chem simulations. Note that sensitivity runs for testing the effect of absorption compare two simulations with fire emissions turned ON, while the sensitivity test for changing the initial emissions and using WRF-Chem show differences of simulations turning smoke emissions on and off. These three sensitivity cases are highlighted in Figure 9 for better comparison.

[Feingold et al., 2005; Johnson et al., 2004]. As seen in Figure 10, absorption makes LCL height shallower for the outbreak on 15 April 2011 while deeper for the other two. This is consistent with difference in the location of smoke between simulations, as for 15 April 2011 cloud fraction tends to increase when the smoke is absorbing while it is reduced for the other two outbreaks (Figure 11, bottom left). Note that for 5 May 2007 the overall effect of smoke on LCL is negligible (Figure 9, top left), thus absorption is counteracting the microphysical effects resulting in no apparent effects. On the other hand, absorption tends to have a negligible effect or increase low-level wind shear and SRH for all outbreaks. The increases are likely due to stabilization of the boundary layer by aerosols on top of it. This is expected for the 15 April 2011 case due to the thick smoke layer on top and also occurs for the 4 May 2003 probably because this outbreak has the largest smoke concentrations and the higher values are at the planetary boundary layer (PBL) height (Figures 7 and 8), so some of this smoke is likely sitting right on top of the PBL generating these effects.

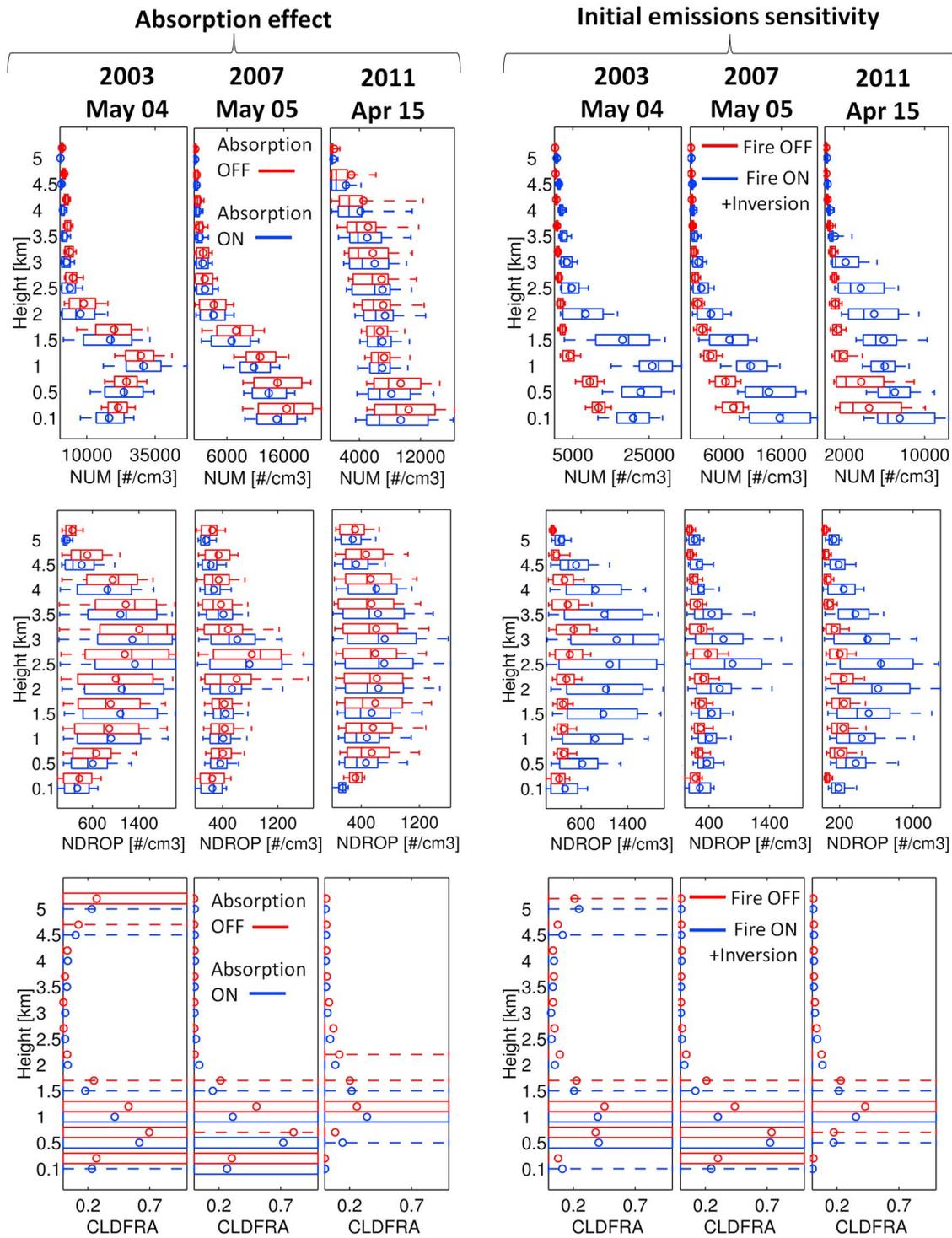


Figure 11. As Figure 7 for two sensitivity cases, turning off (left column) aerosol absorption and performing the inversion with a different set of (right column) initial biomass burning emissions. See details of these simulations on the text. Figure 11 (left column) shows a comparison of two simulations with biomass burning turned on, while Figure 11 (right column) compares simulations with emissions turned on and off.

The smoke effects when using a different set of initial smoke emissions are shown on Figure 10 (middle row) and the vertical profiles can be found on Figure 11 (right column). The sensitivity simulations are performed to assess smoke effects under different smoke loads. By comparing Figure 11 and Figure 7 it can be seen that the resulting aerosol number concentration distributions differ, with 5 May 2007 outbreak showing larger concentration

enhancements due to smoke in the sensitivity simulations and the 4 May 2003 and 15 April 2011 outbreaks showing lower enhancements. The changes in smoke can have a variety of effects. For the 5 May 2007 and 15 April 2011 outbreaks the smoke effects remain similar in the sensitivity simulations (Figure 10) compared to the baseline simulations (Figure 9), i.e., negligible smoke effects for the 5 May 2007 outbreak and small effect size for the 15 April 2011. On the other hand, smoke effects change drastically for the 4 May 2003 outbreak, going from medium-large effect size for the baseline simulation (Figure 9) to small effect size on the sensitivity simulation (Figure 10). This is likely explained due to the lower aerosol loads in the sensitivity simulation which does not generate the cloud fraction enhancement shown in the baseline case (Figure 11 versus 7). However, although there are differences in the magnitude of the smoke effects, the direction of the change remains the same, i.e., smoke generates an increased likelihood of tornado formation and intensity for both cases.

3.5. WRF-AAM Versus WRF-Chem

In this section we compare WRF-AAM and WRF-Chem in terms of the AOD loads and smoke effects.

The constrained emissions are influenced by the ability of the simplified aerosol species in the AAM configuration to represent AOD. Figure 6 (top row) shows a comparison between WRF-AAM and WRF-Chem AOD for one of the outbreaks studied (27 April 2011). The similarity between both simulations is remarkable given the differences in the aerosol model (a full chemistry sectional scheme with eight size bins versus two aerosol species), the optical properties parameterization (a Mie code including a core-shell treatment versus a table lookup approach), and the cloud physics parameterizations (Morrison [Yang *et al.*, 2011, and references therein] versus TE2014 microphysics). Some of the differences between the simulations are due to the way each treats the hygroscopic growth of smoke. In the case of WRF-Chem configured with MOSAIC, aerosol hygroscopic growth is done through the electrolytes in the aerosol mixture, and as organic carbon and black carbon are not considered as electrolytes in this model configuration, there is little hygroscopic growth for smoke. For the AAM configuration the opposite happens, as organic carbon emissions are lumped into the hygroscopic aerosol (hygroscopicity parameter equals to 0.4) and thus undergo significant water uptake. There are large differences in the northwest of the Gulf of Mexico between WRF-AAM and WRF-Chem (Figure 6), which are due to the differences in hygroscopic growth as some model layers show large values of relative humidity (RH) in this region at this time. The localized spikes in the WRF-AAM AOD that are not seen in the WRF-Chem AOD are also due to hygroscopic growth. Thus, we expect to obtain slightly lower emission correction factors when performing the inversion with the AAM than with WRF-Chem due to the larger AOD obtained because of the hygroscopic growth of smoke.

As mentioned previously, nonnegligible smoke effects are found for the 27 April 2011 outbreak when using the AAM configuration. Saide *et al.* [2015b], using WRF-Chem, reported 100–200 m lower LCL, ~2 m/s higher low-level wind shear and ~50 m²/s² higher SRH as a result of smoke interactions with clouds and radiation, which is in the same direction and of similar values to what is found with WRF-AAM for this event (Figure 9). This agreement is again noteworthy given the complex aerosol-cloud-radiation interactions that occur in this system and the way they are parameterized differently in each model configuration (e.g., aerosol optical properties, aerosol size distribution, cloud microphysics, and aerosol activation).

We performed WRF-Chem simulations for the three outbreaks studied in the sensitivity simulations (section 3.4) to further assess similarities and differences with respect to the WRF-AAM configuration. Figures 7 and 12 (top row) show that aerosol number concentrations for simulations with and without fire emissions are consistent within the two systems. Note that the WRF-Chem concentrations are shown only for the accumulation mode, and thus, aerosol number concentrations are much lower compared to WRF-AAM. Although in WRF-Chem total aerosol number is dominated by ultrafine particles from new particle formation, these particles tend to contribute less to cloud condensation nuclei (CCN) than accumulation mode particles, and thus, cloud droplet numbers are higher in simulations with smoke (Figure 12, middle row), which is consistent with WRF-AAM simulations.

As seen in Figure 10 (bottom row), smoke intensifies tornado parameters for all cases when using WRF-Chem with effect size on the small to medium range. This is generally consistent with the WRF-AAM simulations, as with this system the 4 May 2003 and 15 April 2011 outbreaks show intensification with small to large effect size, while 5 May 2007 shows negligible effects (Figure 9, top row). There could be multiple reasons for these discrepancies. For instance, for the 4 May 2003 outbreak the enhancement in cloud droplet number

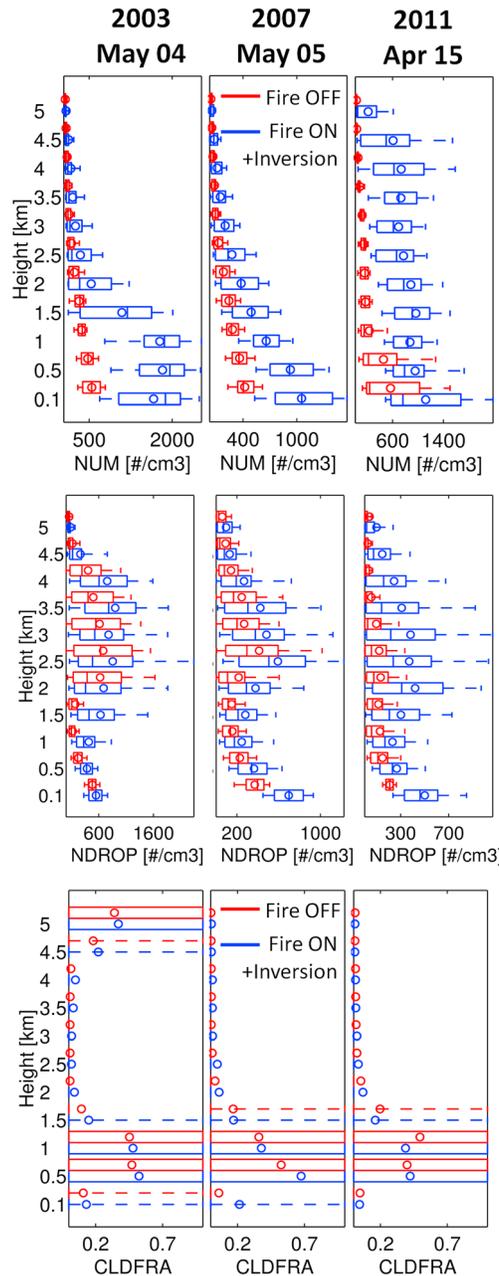


Figure 12. As Figure 7 but for WRF-Chem simulations. Aerosol number corresponds to the accumulations mode (150–625 μm dry diameter).

multiple tornado outbreaks in the U.S. happening during the fire season. To do so, we included emission processes into the WRF-AAM by using WRF-Chem routines that add primary aerosol emissions to the model. Also, we constrained biomass burning emissions for each outbreak studied using satellite-derived AOD and an inverse modeling algorithm. We found a general underestimation of the prior emissions but with large spatial and temporal variations. This is important as studies not using an observational constraint are likely to underestimate the smoke effects, and this cannot be fixed by just applying a global correction factor to emissions.

Using the WRF aerosol-aware-microphysics configuration, we found smoke present in the boundary layer of all outbreaks studied with some cases presenting a thick layer of smoke aloft. Across the various cases studied, we also found a large spread of the smoke effects on environmental conditions, going from negligible impacts to intensifications due to smoke (i.e., a reduction of LCL and increase in low-level wind shear and

concentration due to smoke for WRF-AAM (Figure 7) is larger than for WRF-Chem (Figure 12) which contributes to a larger effect size in tornado parameters for WRF-AAM for this outbreak. The higher cloud droplet number concentration in WRF-AAM could be due to the differences between the aerosol activation parameterizations in WRF-Chem and WRF-AAM as droplet nucleation could get saturated at different aerosol loadings. Aerosol optical properties could also be producing some of these differences, as the smoke is more absorbing on WRF-Chem (SSA of 0.93–0.94 on the region of the 4 May 2003 outbreak) than on WRF-AAM (SSA over 0.96 for RH over 80%); thus, the burn-off of clouds due to absorption found in section 3.4 for the 4 May 2003 outbreak could be occurring more efficiently in WRF-Chem and thus preventing the large enhancement in cloud fraction seen in WRF-AAM. For the 5 May 2007 outbreak the aerosol concentrations are maximum at the surface, which generates a different response in WRF-AAM and WRF-Chem, with WRF-Chem increasing cloud fractions due to smoke at the lower levels (below 750 m) and thus intensifying the tornado parameters where in WRF-AAM only negligible changes are found.

4. Conclusions

In this study we used the WRF modeling system configured with an aerosol-aware microphysics (AAM) parameterization to study impacts of smoke from Central America on multi-

SRH) which are in the range of differences found in the sounding climatology within adjacent supercell classes [Thompson *et al.*, 2003]. Smoke effects on environmental conditions were considerable when there was a distinct layer of low-level clouds (cloud fraction >0.2) and the smoke largely increased aerosol number concentration over background values (threefold to ninefold), producing large cloud droplet concentration increases (twofold to threefold). We found that the presence of a thick layer of smoke above the low-level clouds can contribute to the intensifications of tornado parameters. However, when this layer is not present and smoke concentration at cloud level is large, aerosol absorption can produce cloud burn-off which can deepen the LCL heights, counteracting the microphysical effects. Sensitivity simulations also showed that when changing smoke emissions the smoke effects on tornado parameters were consistent but the magnitudes could vary with the emissions amount.

In this study we only analyzed the effects of biomass burning smoke on environmental conditions that lead to a higher likelihood of tornado formation and intensity. Other studies have proposed alternative mechanisms such as convection invigoration to link smoke with tornadoes [Wang *et al.*, 2009] or suggested that tornado occurrence have a weekly cycle due to anthropogenic activities [Rosenfeld and Bell, 2011]. Recent evidence has shown that fires also present a weekly cycle [Earl *et al.*, 2015]. Given our finding that smoke was present for all outbreaks studied and that can be a major contributor to aerosol number concentrations in all the cases, the role of fires in these other mechanisms needs to be included in future work studying tornadoes during the fire season. Also, given that the Gulf of Mexico is a common source of moisture for severe thunderstorms [Brooks *et al.*, 2003], we expect smoke to be transported along with moisture during the burning season; thus, future work should assess the effects of smoke for severe weather other than tornadoes (e.g., large hail and damaging winds).

Finally, we also compared WRF configured with aerosol-aware-microphysics to WRF-Chem. Although these model configurations use different parameterizations of aerosol optical properties, aerosol size, cloud microphysics, and cloud droplet nucleation, we found their results are generally consistent. In particular, they show comparable AODs over the region affected with smoke, similar smoke transport patterns to the outbreak region, and consistent smoke effects on environmental conditions which can lead to intensification of the outbreak. Thus, the aerosol-cloud-radiation interactions included in WRF-AAM are generally in agreement with more complex models and given the current restrictions on computing power dedicated to these interactions they represent a good choice for the moment. This is encouraging as the AAM is planned to become one of National Centers for Environmental Prediction (NCEP) operational configurations (Rapid Refresh and High-Resolution Rapid Refresh Forecasts). The method to constrain emissions coupled to the WRF-AAM is also computationally efficient as it only requires two simulations (no adjoint and no ensembles); thus, it could be implemented in near-real time applications to constrain highly uncertain emissions such as those from fires and wind-blown dust.

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